

# Equivalence of linear canonical transform domains to fractional Fourier domains and the bicanonical width product: a generalization of the space–bandwidth product

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Linear canonical transforms (LCTs) form a three-parameter family of integral transforms with wide application in optics. We show that LCT domains correspond to scaled fractional Fourier domains and thus to scaled oblique axes in the space–frequency plane. This allows LCT domains to be labeled and ordered by the corresponding fractional order parameter and provides insight into the evolution of light through an optical system modeled by LCTs. If a set of signals is highly confined to finite intervals in two arbitrary LCT domains, the space–frequency (phase space) support is a parallelogram. The number of degrees of freedom of this set of signals is given by the area of this parallelogram, which is equal to the bicanonical width product but usually smaller than the conventional space–bandwidth product. The bicanonical width product, which is a generalization of the space–bandwidth product, can provide a tighter measure of the actual number of degrees of freedom, and allows us to represent and process signals with fewer samples. © 2010 Optical Society of America

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## 1. INTRODUCTION

The conventional space–bandwidth product is of fundamental importance in signal processing and information optics because of its interpretation as the number of degrees of freedom of signals [1–17]. In this paper, we discuss the *bicanonical width product*, which generalizes the space–bandwidth product and which can often provide a tighter measure of the actual number of degrees of freedom.

The definition of the bicanonical width product is based on linear canonical transforms (LCTs), which are a three-parameter family of linear integral transforms [16,18]. The Fourier and fractional Fourier transforms (propagation through quadratic graded-index media), coordinate scaling (imaging), chirp multiplication (passage through a thin lens), and chirp convolution (Fresnel propagation in free space) are some of the special cases of this family of transforms. Because concatenation of LCTs are also LCTs, the family of LCTs can model a broad class of optical systems involving arbitrary combinations of any number of lenses, sections of free space, and sections of graded-index media. These systems belong to the class of quadratic-phase systems, which are also known as *ABCD* systems or lossless first-order optical systems [18–27]. LCTs have also been referred to as generalized Huygens integrals [28], generalized Fresnel transforms [29,30], special affine Fourier transforms [31,32], extended fractional Fourier transforms [33], and Moshinsky–Quesne transforms [18], among other things, and have found use in image filtering [34]. In this paper we show that LCT

domains are essentially equivalent to fractional Fourier domains.

We will provide motivation, interpretation, and an overview of the main ideas and results in Section 2, which will then be substantiated in Sections 3–5. The definition and properties of LCTs will be briefly reviewed in Section 3. In Section 4 we establish the equivalence of LCT domains to fractional Fourier domains. The relationships between the space-frequency support, the bicanonical width product, and the number of degrees of freedom is the subject of Section 5. We conclude in Section 6. This work is based on [35].

## 2. OVERVIEW

### A. Equivalence of Linear Canonical Transform Domains to Fractional Fourier Domains

One of the most important concepts in Fourier analysis is the concept of the frequency (or Fourier) domain. This domain is understood to be a space where the frequency representation of the signal lives. Since the fractional Fourier transform (FRT) has the effect of rotating the space-frequency (phase space) representation of a signal, *fractional Fourier domains* correspond to oblique axes in the space-frequency plane (phase space) [16,36]. By analogy with the concept of fractional Fourier domains, the term *linear canonical domain* has been used in several papers to refer to the domain of the LCT representation of a signal [37–46]. However, it has not been clear what or where these domains are. In other words, while the effect

of an LCT on the space-frequency representation of a signal is well understood as a linear geometrical distortion, it has not been obvious how members of the three-parameter family of LCT domains are related to the space-frequency plane, or how we should visualize them.

One of our contributions is to explicitly relate LCT domains to the space-frequency plane. We will show that *each LCT domain corresponds to a scaled FRT domain and thus a scaled oblique axis in the space-frequency plane* [35]. Based on this many-to-one association of LCTs with FRTs, LCT domains can be labeled and monotonically ordered by an associated fractional order parameter, instead of their usual three parameters, which do not directly lend themselves to a natural ordering.

This has important implications for optical systems modeled by LCTs (such as those consisting of arbitrary concatenations of lenses and sections of free space). The optical amplitude distribution at any plane along the optical axis can be expressed as a three-parameter LCT of the input. Being able to associate a single monotonically increasing parameter with each location along the optical axis offers a vastly more transparent understanding of the evolution of the optical amplitude distribution, as opposed to imagining that the light goes through an unidentified and unsequenced series of three-parameter domains whose whereabouts we cannot visualize. This interpretation of LCT domains also ties together the LCT description of such optical systems with descriptions viewing propagation as an act of continual fractional Fourier transformation [47,48].

The association between LCT and FRT domains is based on expressing the LCT as a chirp multiplied and scaled FRT, which is a special case of the widely known Iwasawa decomposition [24,25,49]. Multiplication with a function is not usually considered to be an operation that changes the domain of a signal, and scaling of the axis is a relatively trivial modification of a domain. Therefore, LCTs of a signal that are associated with the same FRT order parameter essentially live in the same domain (the FRT domain with that order). We will speak of LCT domains sharing the same order parameter as essentially equivalent domains. If a signal is confined to a finite interval in a certain LCT domain, it will also be confined to a finite interval in all essentially equivalent domains.

## B. Generalization of the Space-Bandwidth Product: The Bicanonical Width Product

### 1. Space-Bandwidth Product

Let us first recall the construction of the concept of the conventional space-bandwidth product. Consider a family of signals whose members are approximately confined to an interval of length  $\Delta u$  in the space domain and to an interval of length  $\Delta \mu$  in the frequency domain in the sense that a large percentage of the signal energy is confined to these intervals. The space-bandwidth product  $N$  is defined [12,16] as

$$N \equiv \Delta u \Delta \mu, \quad (1)$$

and is always greater than or equal to unity because of the uncertainty relation.

The conventional space-bandwidth product is the minimum number of samples required to uniquely identify a signal out of all possible signals whose energies are approximately confined to space and frequency intervals of length  $\Delta u$  and  $\Delta \mu$ . This argument is based on the Shannon–Nyquist sampling theorem, which requires that the spacing between samples not be greater than  $\delta u = 1/\Delta \mu$ , so that the minimum number of samples over the extent  $\Delta u$  is given by  $\Delta u/\delta u = \Delta u \Delta \mu$ . Alternatively, if we sample the signal in the frequency domain, the spacing between samples should not be greater than  $\delta \mu = 1/\Delta u$ , so that the minimum number of samples over the extent  $\Delta \mu$  is given by  $\Delta \mu/\delta \mu = \Delta u \Delta \mu$ . The minimum number of samples needed to fully characterize an approximately space- and band-limited signal can also be interpreted as the number of degrees of freedom of the set of signals. This number of samples turns out to be the same whether counted in the space or frequency domain, and is given by the space-bandwidth product.

When the approximate space and frequency extents are specified as above, this amounts to assuming that most of the energy of the signal is confined to a  $\Delta u \times \Delta \mu$  rectangular region in the space-frequency plane, perpendicular to the space-frequency axes (Fig. 1). In this case, the number of degrees of freedom and the space-bandwidth product are both equal to the area of this rectangular region. More generally, the number of degrees of freedom is given by the area of the space-frequency support (phase space support), regardless of its shape [14,16]. (The space-frequency support can be roughly defined as the region in the space-frequency plane in which a large percentage of the total energy is confined.) When the space-frequency support is not a rectangle perpendicular to the axes, the actual number of degrees of freedom will be smaller than the space-bandwidth product of the signal [14,16].

The space-bandwidth product is a notion originating from the simultaneous specification of the space and frequency extents. Although this product is commonly seen as an intrinsic property, it is in fact a notion that is specific to the Fourier transform and the frequency domain. It is also possible to specify the extents in other FRT or LCT domains. The set of signals thus specified will in general exhibit a nonrectangular space-frequency support. We will show that when two such extents are specified, the support will be a parallelogram. In all cases, the area of the support will correspond to the number of degrees of freedom of the set of signals thus defined. If we insist on characterizing this set of signals with conventional space

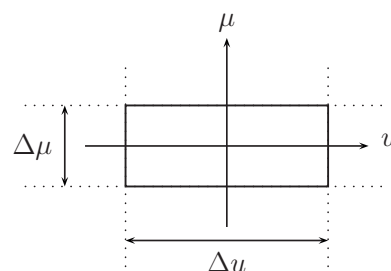


Fig. 1. Rectangular space-frequency support with area equal to the space-bandwidth product  $\Delta u \Delta \mu$ .

and frequency extents, the space–bandwidth product will overstate the number of degrees of freedom.

Obviously, specifying a finite extent in a single LCT domain does not define a family of signals with a finite number of degrees of freedom, just as specifying a finite extent in only one of the conventional space or frequency domains does not. However, specifying finite extents in two distinct LCT domains allows us to define a family of signals with a finite number of degrees of freedom. The number of degrees of freedom will depend on both the specified LCT domains and the extents in those domains.

## 2. Bicanonical Width Product

We now define the *space–canonical width product*, which gives the number of degrees of freedom of signals that are approximately confined to a finite interval  $\Delta u$  in the conventional space domain and to a finite interval  $\Delta u_{\mathbf{T}}$  in some other LCT domain [35,50]:

$$N \equiv \Delta u \Delta u_{\mathbf{T}} |\beta|. \quad (2)$$

This is always greater than or equal to unity because of the uncertainty relation for LCTs [16,18,40,45]. Here  $\mathbf{T}$  represents the three parameters of the LCT, which will be explicitly defined in Section 3. Here  $\beta$ , which can assume any real value, is one of the three parameters of the LCT in question. The space–canonical width product constitutes a generalization of the space–bandwidth product, and reduces to it when the LCT reduces to an ordinary Fourier transform, upon which  $\Delta u_{\mathbf{T}}$  reduces to  $\Delta \mu$  and  $\beta=1$ .

In the above, one of the two domains was chosen to be the conventional space domain. More generally, the two LCT domains can both be arbitrarily chosen. In this case, we use the more general term *bicanonical width product* to refer to the product [35,50]

$$N \equiv \Delta u_{\mathbf{T}_1} \Delta u_{\mathbf{T}_2} |\beta_{1,2}|, \quad (3)$$

where  $\Delta u_{\mathbf{T}_1}$  and  $\Delta u_{\mathbf{T}_2}$  are the extents of the signal in two arbitrary LCT domains and  $\beta_{1,2}$  is the parameter of the LCT between these two domains (the LCT that transforms the signal from the first LCT domain to the second). If  $\beta=\infty$  in Eq. (2) or  $\beta_{1,2}=\infty$  in Eq. (3), then the product  $N$  will not be finite, and the number of degrees of freedom will not be finite. We shall see in Section 3 that when this parameter is infinity, the two domains are related to each other simply by a scaling or chirp multiplication operation. But as discussed before, domains related by such operations are essentially equivalent. Thus, specification of the extent in two such domains does not constrain the family of signals more than the specification of the extent in only one domain, which, as noted, is not sufficient to make the number of degrees of freedom finite.

The space–canonical width product is the minimum number of samples required to uniquely identify a signal out of all possible signals whose energies are approximately confined to a space interval of  $\Delta u$  and a particular LCT interval of  $\Delta u_{\mathbf{T}}$ . This number of samples can be used to reconstruct the signal. This can be justified by the use of the LCT sampling theorem [38,51,52], which is a generalization of the FRT sampling theorem [53–58]. According to the LCT sampling theorem, the space-domain sam-

pling interval for a signal that has finite extent  $\Delta u_{\mathbf{T}}$  in a particular LCT domain should not be larger than  $\delta u = 1/(|\beta| \Delta u_{\mathbf{T}})$ . If we sample the space-domain signal at this rate, the total number of samples over the extent  $\Delta u$  will be given by  $\Delta u / \delta u = \Delta u \Delta u_{\mathbf{T}} |\beta|$ , which is precisely equal to the space–canonical width product. Alternatively, if we sample in the LCT domain, the sampling interval should not be larger than  $\delta u_{\mathbf{T}} = 1/(|\beta| \Delta u)$ . Sampling at this rate, the total number of samples over the extent  $\Delta u_{\mathbf{T}}$  is given by  $\Delta u_{\mathbf{T}} / \delta u_{\mathbf{T}} = \Delta u \Delta u_{\mathbf{T}} |\beta|$ , which once again is the space–canonical width product.

The derivation above can be easily replicated for the more general bicanonical width product defined in Eq. (3). Therefore, *the bicanonical width product can also be interpreted as the minimum number of samples required to uniquely identify a signal out of all possible signals whose energies are approximately confined to finite intervals in the two specified LCT domains, and therefore as the number of degrees of freedom of this set of signals.*

## 3. Space–Frequency Support and Degrees of Freedom

Finite extent constraints in the space and frequency domain are well-known to translate to vertical and horizontal corridor constraints in the space–frequency plane. (With the term “corridor” we are referring to an infinite strip in the space–frequency plane perpendicular to a particular domain.) This result can be precisely stated in terms of the Wigner distribution, one of the most frequently used space–frequency representations in optics [59–63]. It is not difficult to extend this result to FRT domain constraints, which translate to oblique corridor constraints in the space–frequency plane. However, without the equivalence of LCT domains to scaled FRT domains that we present, the effect of LCT domain constraints in the space–frequency plane has not been clear. With this equivalence, it becomes possible to see that LCT domain constraints also translate to oblique corridor constraints. Therefore, *restricting the signal to finite intervals in two distinct LCT domains implies that the space–frequency support is a parallelogram (see Fig. 2 below); moreover, the area of the parallelogram is equal to the bicanonical width product [35].* (Indeed, the bicanonical width product can be interpreted as a formula for the area of the parallelogram.) This provides further justification for interpreting the bicanonical width product as the number of degrees of freedom of LCT-limited signals.

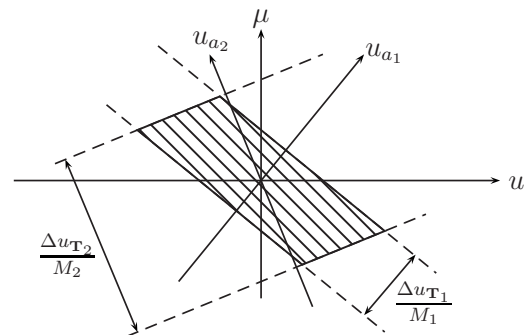


Fig. 2. Space–frequency support when finite extents are specified in two LCT domains. The area of the parallelogram is equal to  $\Delta u_{\mathbf{T}_1} \Delta u_{\mathbf{T}_2} |\beta_{1,2}|$ . This figure will be revisited in Section 5.



#### 4. Efficient Signal Representation

There is no reason to think that families of signals encountered in practice will necessarily uniformly fall into a rectangular region in the ordinary space–frequency plane. As well understood, when the family of signals does not have a rectangular space–frequency support perpendicular to the space–frequency axes, the space–bandwidth product overstates the actual number of degrees of freedom. That is, we can represent these signals with a number of samples less than the space–bandwidth product. In some applications, where the underlying physics involves LCT type integrals (as is the case with many wave propagation problems and optical systems), parallelograms may be excellently, if not perfectly, tailored to the true space–frequency supports.

Even when this is not the case, it may frequently be possible to more tightly enclose the actual space–frequency support with a parallelogram than with a rectangle perpendicular to the axes, or indeed with any rectangle. Then, the bicanonical width product may better represent the actual number of degrees of freedom, and we will be able to represent the signals with fewer samples [35].

Even when confronted with a space–frequency support of arbitrary shape, it is quite common to assume the number of degrees of freedom to be equal to the space–bandwidth product and work with that many samples, without regard to the shape of the space–frequency support. In reality, this is an inefficient worst-case approach, which encloses the arbitrary shape within a rectangle perpendicular to the axes, and thus which overstates the number of degrees of freedom. Why is this approach so common? Because in general there is no easy and general way of handling arbitrary irregular space–frequency supports. There is no guarantee that such signals can be represented with space domain or other linear unitary transform domain samples with the number of samples being equal to the number of degrees of freedom. In general, it is necessary to find a custom basis in which the signals can be represented with a number of coefficients equal to the number of degrees of freedom, and it may not always be easy to find such a convenient basis.

While our approach does not fully solve this problem, it offers a general approach allowing the representation and processing of signals with a number of samples closer to their true number of degrees of freedom. *Having the freedom to enclose a given region with an arbitrary parallelogram is much more flexible than being restricted to enclosing the region with a rectangle perpendicular to the axes, or indeed with any rectangle, and in general will allow a smaller area and number of samples to be achieved* [35]. Once a parallelogram minimally enclosing the true space–frequency support is found (which gives us two LCT domains in which the signals are confined to the extents  $\Delta u_{T_1}$  and  $\Delta u_{T_2}$ ), the signal can be represented with  $\Delta u_{T_1} \Delta u_{T_2} |\beta_{1,2}|$  uniform samples in these LCT domains.

Therefore, the problem of finding the smallest number of samples required to represent signals with an arbitrary space–frequency support reduces to a simple geometrical problem [35]. We simply need to find the smallest parallelogram enclosing the space–frequency support. Then the area of the parallelogram, which is equal to the bica-

nonical width product, will give us the minimum number of samples required to represent the signal using the LCT sampling theorem. The continuous signals can be recovered using the LCT interpolation formula [44,46].

#### 5. Relation to the Discrete LCT

It is worth tying our development to a recent result [50] that showed that *if the number of samples  $N$  is chosen to be at least equal to the bicanonical width product, the discrete LCT (DLCT) can be used to obtain a good approximation to the continuous LCT*, limited only by the fundamental fact that a signal cannot have *strictly* finite extent in more than one domain [50,64]. The exact relation between the discrete and continuous LCT derived in [50] precisely shows the approximation involved and demonstrates how the approximation improves with increasing  $N$ .

Because this exact relation generalizes the corresponding relation for Fourier transforms [65], the DLCT defined in [66] and discussed in [50,64,67,68] approximates the continuous LCT in the same sense that the DFT approximates the continuous Fourier transform. This elegant and natural formulation of discrete LCT computation mirrors several of the concepts we have been discussing. Rather than assuming that the space and frequency extents of the signals are specified in the conventional manner and determining the number of samples from the standard Nyquist–Shannon sampling theorem as in [69–75], in this formulation the extents are specified in the input and output LCT domains, and the number of samples are determined from the LCT sampling theorem [50]. The minimum number of samples required for computation is given by the bicanonical width product and the DLCT defined in [66] works with this number of samples without requiring any interpolation or oversampling. On the other hand, sampling the input and output conventionally as in the above cited papers usually leads to a greater number of samples and sometimes requires different numbers of samples for the input and output signals. Finally we note that the DLCT in [50,64,66–68] can be efficiently computed in  $\mathcal{O}(N \log N)$  time by successively performing a chirp multiplication, a fast Fourier transform (FFT), and a second chirp multiplication [50,56,66] in a straightforward manner without the need for sophisticated algorithms or space–frequency support tracking.

Fast computation, a well-defined relationship to the continuous LCT, and unitarity make this definition of the DLCT an important candidate for being a widely accepted definition of the discrete version of the LCT.

### 3. LINEAR CANONICAL TRANSFORMS

Optical systems involving thin lenses, sections of free space in the Fresnel approximation, sections of quadratic graded-index media, and arbitrary combinations of any number of these are referred to as first-order optical systems or quadratic-phase systems. Mathematically, such systems can be modeled as linear canonical transforms (LCTs). The output light field  $f_T(u)$  of a quadratic-phase system is related to its input field  $f(u)$  through [16,18]

$$f_{\mathbf{T}}(u) \equiv (\mathcal{C}_{\mathbf{T}}f)(u) \equiv \int_{-\infty}^{\infty} C_{\mathbf{T}}(u, u') f(u') du',$$

$$C_{\mathbf{T}}(u, u') \equiv \sqrt{\frac{1}{B}} e^{-i\pi/4} e^{i\pi \left( \frac{D}{B} u^2 - 2\frac{1}{B} u u' + \frac{A}{B} u'^2 \right)}, \quad (4)$$

for  $B \neq 0$ , where  $\mathcal{C}_{\mathbf{T}}$  is the unitary LCT operator with parameter matrix  $\mathbf{T} = [A \ B; C \ D]$  with  $AD - BC = 1$ . In the trivial case  $B = 0$ , the LCT is defined simply as  $f_{\mathbf{T}}(u) \equiv \sqrt{D} \exp[i\pi C D u^2] f(Du)$ . Sometimes the three real parameters  $\alpha = D/B$ ,  $\beta = 1/B$ ,  $\gamma = A/B$  are used instead of the unit-determinant matrix  $\mathbf{T}$  whose elements are  $A, B, C, D$ . (One of the four matrix parameters is redundant because of the unit-determinant condition). These two sets of parameters are equivalent, and either set of parameters can be obtained from the other [16,18]:

$$\mathbf{T} = \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} \gamma/\beta & 1/\beta \\ -\beta + \alpha\gamma/\beta & \alpha/\beta \end{bmatrix}. \quad (5)$$

The transform matrix  $\mathbf{T}$  is useful in the analysis of optical systems because if several systems are cascaded, the overall system matrix can be found by multiplying the corresponding matrices.

The LCT family includes the Fourier and fractional Fourier transforms, coordinate scaling, and chirp multiplication and convolution operations as its special cases.

The  $a$ th-order fractional Fourier transform (FRT) [16] of a function  $f(u)$ , denoted by  $f_a(u)$ , is most commonly defined as

$$f_a(u) \equiv (\mathcal{F}^a f)(u) \equiv \int_{-\infty}^{\infty} K_a(u, u') f(u') du',$$

$$K_a(u, u') \equiv A_{\phi} e^{i\pi(\cot\phi u^2 - 2\csc\phi u u' + \cot\phi u'^2)},$$

$$A_{\phi} = \sqrt{1 - i \cot \phi}, \quad \phi = a\pi/2 \quad (6)$$

when  $a \neq 2j$ ,  $K_a(u, u') = \delta(u - u')$  when  $a = 4j$ , and  $K_a(u, u') = \delta(u + u')$  when  $a = 4j \pm 2$ , where  $j$  is an integer. The FRT is a special case of the LCT with matrix

$$\mathbf{F}^a = \begin{bmatrix} \cos \phi & \sin \phi \\ -\sin \phi & \cos \phi \end{bmatrix}, \quad (7)$$

differing only by an inconsequential factor:  $\mathcal{C}_{\mathbf{F}^a} f(u) = e^{-ia\pi/4} \mathcal{F}^a f(u)$  [16,18].

Another special case of the LCT is multiplication with a chirp function of the form  $\exp[-i\pi q u^2]$ , which corresponds to a thin lens in optics. The corresponding LCT matrix is given by

$$\mathbf{Q}_q = \begin{bmatrix} 1 & 0 \\ -q & 1 \end{bmatrix}. \quad (8)$$

Propagation through a section of free space in the Fresnel approximation is equivalent to convolution with a chirp function of the form  $e^{-i\pi/4} \sqrt{1/r} \exp[i\pi u^2/r]$ . The corresponding LCT matrix is given by

$$\mathbf{R}_r = \begin{bmatrix} 1 & r \\ 0 & 1 \end{bmatrix}. \quad (9)$$

Optical imaging can be modeled as a scaling operation, which maps a function  $f(u)$  into  $\sqrt{1/M} f(u/M)$  with  $M > 0$ . The transformation matrix is

$$\mathbf{M}_M = \begin{bmatrix} M & 0 \\ 0 & 1/M \end{bmatrix}. \quad (10)$$

An arbitrary LCT can be decomposed into a FRT followed by scaling followed by chirp multiplication [16,71]:

$$\mathbf{T} = \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ -q & 1 \end{bmatrix} \begin{bmatrix} M & 0 \\ 0 & 1/M \end{bmatrix} \begin{bmatrix} \cos \phi & \sin \phi \\ -\sin \phi & \cos \phi \end{bmatrix}. \quad (11)$$

Here,  $q$  is the chirp multiplication parameter,  $M > 0$  is the scaling factor, and  $\phi = a\pi/2$ , where  $a$  is the order of the FRT. The decomposition can be written more explicitly in terms of the LCT and FRT domain representations of the signal as

$$f_{\mathbf{T}}(u) = \exp[-i\pi q u^2] \sqrt{\frac{1}{M}} f_a\left(\frac{u}{M}\right). \quad (12)$$

This decomposition is a special case of the Iwasawa decomposition [24,25,49]. (For a discussion of the implications of this decomposition to the propagation of light through first-order optical systems, see [47].) As we will see, by appropriately choosing the three parameters  $a, M, q$ , the above equality can be satisfied for any  $ABCD$  matrix. Solving for  $a, M, q$  in Eq. (11), we obtain the decomposition parameters in terms of the matrix entries  $A, B, C, D$ :

$$a = \begin{cases} \frac{2}{\pi} \arctan\left(\frac{B}{A}\right), & \text{if } A \geq 0 \\ \frac{2}{\pi} \arctan\left(\frac{B}{A}\right) + 2, & \text{if } A < 0 \end{cases}, \quad (13)$$

$$M = \sqrt{A^2 + B^2}, \quad (14)$$

$$q = \begin{cases} -\frac{C}{A} - \frac{B/A}{A^2 + B^2}, & \text{if } A \neq 0 \\ -\frac{D}{B}, & \text{if } A = 0 \end{cases}. \quad (15)$$

The range of the arccotangent lies in  $(-\pi/2, \pi/2]$ .

The Wigner distribution (WD) of a signal  $W_f(u, \mu)$  is a phase-space distribution that gives the distribution of signal energy over space and frequency, and is defined as [16,76]

$$W_f(u, \mu) = \int_{-\infty}^{\infty} f(u + u'/2) f^*(u - u'/2) e^{-i2\pi\mu u'} du'. \quad (16)$$

The WD of  $f_{\mathbf{T}}(u)$  can be related to the WD of  $f(u)$  by a linear distortion [16]:

$$W_{f_T}(u, \mu) = W_f(Du - B\mu, -Cu + A\mu). \quad (17)$$

The Jacobian of this coordinate transformation is equal to the determinant of the matrix  $\mathbf{T}$ , which is unity. Therefore, this coordinate transformation will geometrically distort the support region of the WD, but the support area will remain unchanged.

#### 4. RELATION BETWEEN FRACTIONAL FOURIER DOMAINS AND LINEAR CANONICAL DOMAINS

As is well-known, fractional Fourier domains correspond to oblique axes in the space–frequency plane [36]. The effect of  $a$ th-order fractional Fourier transformation on the Wigner distribution of a signal is to rotate the Wigner distribution by an angle  $\phi = a\pi/2$  [36,77,78]:

$$W_{f_a}(u, \mu) = W_f(u \cos \phi - \mu \sin \phi, u \sin \phi + \mu \cos \phi). \quad (18)$$

The Radon transform operator  $\mathcal{RDN}_\phi$ , which takes the integral projection of the Wigner distribution of  $f(u)$  onto an axis making an angle  $\phi$  with the  $u$  axis, can be used to restate this property in the following manner [16]:

$$\{\mathcal{RDN}_\phi[W_f(u, \mu)]\}(u_a) = |f_a(u_a)|^2, \quad (19)$$

where  $u_a$  denotes the axis making angle  $\phi = a\pi/2$  with the  $u$  axis. That is, projection of the Wigner distribution of  $f(u)$  onto the  $u_a$  axis gives  $|f_a(u_a)|^2$ , the squared magnitude of the  $a$ th-order FRT of the function. Hence, the projection axis  $u_a$  can be referred to as the  $a$ th-order fractional Fourier domain (see Fig. 3) [36,77]. The space and frequency domains are merely special cases of the continuum of fractional Fourier domains.

Recently, there has been considerable interest in generalizing the fractional Fourier transform and its properties to linear canonical transforms. In analogy with fractional Fourier domains, the term *LCT domain* has been used to refer to the domain of the LCT representation of a signal [37–40,44,45,79]. However, unlike fractional Fourier domains, which are recognized as oblique axes in the space–frequency plane [16,18,36], it has not been well understood where these LCT domains exist and how they are related to the space–frequency plane. Moreover, LCT domains are characterized by three independent parameters. Therefore, LCT domains populate a three-parameter space, which makes them hard to visualize.

We first introduce the concept of *essentially equivalent domains* by using the Iwasawa decomposition. As given in

Eq. (12), any arbitrary LCT can be expressed as a chirp multiplied and scaled FRT. Thus, in order to compute the LCT of a signal, we can first compute the  $a$ th-order FRT of the signal, which transforms the signal to the  $a$ th-order fractional Fourier domain. Second, we scale the transformed signal. Because scaling is a relatively trivial operation, we need not interpret it as changing the domain of the signal, but merely a scaling of the coordinate axis in the same domain. Finally, we multiply the resulting signal with a chirp to obtain the LCT. Multiplication with a function is not considered an operation that transforms a signal to another domain, but one that alters the signal in the same domain. (For instance, when we multiply the Fourier transform of a function with a mask, the result is considered to remain in the frequency domain.) Therefore, only the FRT part of the LCT operation corresponds to a genuine domain change, and the linear canonical transformed signal essentially lives in a scaled fractional Fourier domain. In other words, *LCT domains are essentially equivalent to scaled fractional Fourier domains*. Despite their three parameters, LCT domains do not constitute a richer family of domains than FRT domains. More explicitly:

In the space–frequency plane, the LCT domain with parameter matrix  $[A \ B; C \ D]$  corresponds to a scaled oblique axis making angle  $\arctan(B/A)$  with the  $u$  axis (or equivalently having slope  $B/A$ ) and with scale parameter  $M = \sqrt{A^2 + B^2}$ .

Note that any LCT domain is completely characterized by the two parameters  $A$  and  $B$  (or equivalently by  $a$  and  $M$ , or by  $\gamma$  and  $\beta$ ) instead of all three of its parameters.

Observe that LCTs with the same value of  $B/A$  (or equivalently the same value of  $\gamma$ ) will have the same value of  $a$  in the decomposition in Eq. (13), and therefore will be associated with the same FRT domain. We refer to such LCT domains, as well as their associated FRT domain, as *essentially equivalent domains*. Note that if a signal has a compact support in a certain LCT domain, then the signal will also have compact support in all essentially equivalent domains.

The concept of essentially equivalent domains we introduce allows many earlier observations and results to be seen in a new light, making them almost obvious or more transparent. For instance, it has been stated that if a particular LCT of a signal is band limited, then another LCT of the signal cannot be band limited unless  $B_1/A_1 = B_2/A_2$  [40]. Since we recognize the domains associated with two LCTs satisfying this relation to be essentially equivalent, this result becomes obvious. Although we will not further elaborate, other results regarding the compactness/band limitedness of different LCTs of a signal [80] can be likewise easily understood in terms of the concept of essentially equivalent domains. As a final example, we consider the LCT sampling theorem [38,51,52], according to which if the LCT of a signal has finite extent  $\Delta u_T$ , then we should sample it with spacing  $\Delta u \leq |B|/\Delta u_T$ . Such a sampling scheme collapses when  $B=0$ . It is easy to understand why if we note that  $B=0$  implies that the LCT domain in question is essentially equivalent to the  $a=0$ th FRT domain; that is, the domain in which

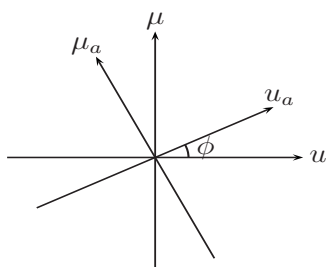


Fig. 3.  $a$ th-order fractional Fourier domain.

the signal is specified to have finite extent is essentially equivalent to the domain in which we are attempting to sample the signal.

Let us now optically interpret the equivalence of LCT domains to FRT domains. Consider a signal that passes through an arbitrary quadratic-phase system. Since the light field at any plane within the system is related to the input field through an LCT, the signal will incrementally be transformed through different LCT domains. Because the three parameters of the consequential LCT domains are not sequenced, it is not easy to give any interpretation or visualize the nature of the transformation of the optical field. However, if we think of the LCT domains as being equivalent to scaled FRT domains, it becomes possible to interpret every location along the propagation axis as a FRT domain of specific order. Moreover, it has been shown that if we take the fractional order  $a$  to be equal to zero at the input of the system, then  $a$  monotonically increases as a function of the distance along the optical axis [47,48]. In other words, propagation through a quadratic-phase system can be understood as passage through a continuum of scaled FRT domains of monotonically increasing order, instead of passage through an unsequenced plethora of LCT domains.

(To see that the FRT parameter  $a$  is monotonically increasing along the  $z$  axis, observe from Eq. (13) that  $a \propto \arctan(B/A)$ , so that  $a$  increases with  $B/A$ . Passage through a lens involves multiplication with the matrix given in Eq. (8), which does not change  $B/A$ . Passage through an incremental section of free space involves multiplication with the matrix given in Eq. (9), which always results in a positive increment in  $a$ . This is because  $r$  is proportional to the distance of propagation, and the derivative of the new value of  $B/A$  with respect to  $r$  is always positive, which implies that  $B/A$  always increases with  $r$ . A similar argument is possible for quadratic graded-index media. A more precise development may be found in [47,48].)

The distribution of light is continually fractional Fourier transformed through fractional domains of increasing order, which we know are oblique axes in the space-frequency plane. This understanding of quadratic-phase systems yields much more insight into the nature of how light is transformed as it propagates through such a system, as opposed to thinking of it in terms of going through a series of unsequenced LCT domains whose whereabouts we cannot visualize.

As a final remark, we note that the relation in Eq. (19) can be rewritten for the LCT of a signal as

$$\frac{1}{M} \{ \mathcal{RDN}_\phi[W_f(u, \mu)] \} \left( \frac{u_T}{M} \right) = |f_T(u_T)|^2, \quad (20)$$

by using Eqs. (19) and (12), where  $u_T$  denotes the axis making angle  $\phi = \arctan(B/A)$  with the  $u$  axis and scaled by  $M$ . This is another way of interpreting scaled oblique axes in the space-frequency plane as the LCT domain with parameter  $\mathbf{T}$ .

## 5. SPACE-FREQUENCY SUPPORT AND DEGREES OF FREEDOM

Having established the relationship of LCT domains to the space-frequency plane, we can now investigate the

space-frequency support of the set of signals whose extents are approximately limited in two LCT domains.

Let us consider a set of signals whose members are approximately confined to the intervals  $[-\Delta u_{T_1}/2, \Delta u_{T_1}/2]$  and  $[-\Delta u_{T_2}/2, \Delta u_{T_2}/2]$  in two given LCT domains,  $u_{T_1}$  and  $u_{T_2}$ . We want to investigate the space-frequency support of this set of signals. Since LCT domains are equivalent to scaled fractional Fourier domains, each finite interval in an LCT domain will correspond to a scaled interval in the equivalent FRT domain. To see this explicitly, we again refer to Eq. (12), which implies that if  $f_T(u)$  is confined to an interval of length  $\Delta u_T$ , so is  $f_a(u/M)$ . Therefore, the extent of  $f_a(u)$  in the equivalent  $a$ th-order FRT domain is  $\Delta u_T/M$ . Thus, the set of signals in question is approximately limited to an extent of  $\Delta u_{T_1}/M_1$  in the  $a_1$ th-order FRT domain, and an extent of  $\Delta u_{T_2}/M_2$  in the  $a_2$ th-order FRT domain, where  $a_1, a_2$  and  $M_1, M_2$  are related to  $\mathbf{T}_1, \mathbf{T}_2$  through Eqs. (13) and (14).

It is well known that if the space-, frequency-, or FRT-domain representation of a signal is identically zero (or negligible) outside a certain interval, so is its Wigner distribution [16,59]. As a direct consequence of this fact, the Wigner distribution of our set of signals is confined to corridors of width  $\Delta u_{T_1}/M_1$  and  $\Delta u_{T_2}/M_2$  in the directions orthogonal to  $u_{a_1}$  and  $u_{a_2}$ , respectively. (With the term ‘‘corridor’’ we are referring to an infinite strip in the space-frequency plane perpendicular to the  $u_a$  axis.) The parallelogram defined by the intersection of these two corridors gives us the space-frequency support of the signals (see Fig. 2). The area of the parallelogram is equal to the bicanonical width product of the set of signals in question. This result will be stated as a theorem.

**Theorem 1.** *Consider a set of signals whose members are approximately confined to finite extents  $\Delta u_{T_1}$  and  $\Delta u_{T_2}$  in the two LCT domains  $u_{T_1}$  and  $u_{T_2}$ , respectively. Let  $\beta_{1,2}$  denote the  $\beta$  parameter of the LCT that transforms signals from the first LCT domain to the second. Then, the area of the space-frequency support of these signals, which is given by a parallelogram defined by these extents (Fig. 2), is equal to the bicanonical width product  $\Delta u_{T_1} \Delta u_{T_2} |\beta_{1,2}|$  of the set of signals.*

*Proof.* The two heights of the parallelogram defined by the extents  $\Delta u_{T_1}$  and  $\Delta u_{T_2}$  are  $\Delta u_{T_1}/M_1$  and  $\Delta u_{T_2}/M_2$ , corresponding to the widths of the corridors. Moreover, the angle between the corridors is  $\phi_2 - \phi_1$ . Then, the area of the parallelogram is

$$\text{Area} = \frac{\Delta u_{T_1}}{M_1} \frac{\Delta u_{T_2}}{M_2} |\csc(\phi_2 - \phi_1)| \quad (21)$$

$$= \frac{\Delta u_{T_1} \Delta u_{T_2}}{M_1 M_2 |\sin \phi_2 \cos \phi_1 - \cos \phi_2 \sin \phi_1|} \quad (22)$$

$$= \frac{\Delta u_{T_1} \Delta u_{T_2}}{|A_1 B_2 - B_1 A_2|} \quad (23)$$

$$= \Delta u_{T_1} \Delta u_{T_2} \frac{|\beta_1 \beta_2|}{|\gamma_1 - \gamma_2|} \quad (24)$$



$$= \Delta u_{T_1} \Delta u_{T_2} |\beta_{1,2}|, \quad (25)$$

where the third and fourth equalities follow from Eqs. (11) and (5), respectively. The final result can be obtained from the matrix corresponding to the LCT operation between these two domains, which is given by  $T_2 T_1^{-1}$ .  $\square$

Since the number of degrees of freedom of a set of signals is given by the area of their space–frequency support, the theorem provides further justification for interpreting the bicanonical width product as the number of degrees of freedom of LCT-limited signals.

Let us now for a moment focus on the space–canonical width product; we assume a finite extent has been specified in the space domain and in some other LCT domain. The corresponding space–frequency support is shown in Fig. 4 (left). If we transform to precisely the same LCT domain in which the extent has been specified, the new space–frequency support becomes as shown in Fig. 4 (right). Here  $M$  and  $M'$  are the scaling parameters associated with the LCT and inverse LCT operations, respectively. Note that in both parts of the figure, the support is bounded by a vertical corridor, perpendicular to the space domain in the left part and to the LCT domain in the right. We are not surprised that the transformed support is again a parallelogram, since the linear geometric distortion imparted by an LCT always maps a parallelogram to another parallelogram. Moreover, the areas of both parallelograms are equal to each other and given by the bicanonical width product  $\Delta u \Delta u_T |\beta|$ , so that the number of degrees of freedom also remains the same after the LCT. This does not surprise us either, since LCTs are known not to change the support area in phase space. We may summarize by stating that *the bicanonical width product is a measure for the number of degrees of freedom of signals that is invariant under linear canonical transformation* [35]. The fact that LCTs model an important family of optical systems makes the bicanonical width product a suitable invariant measure for the number of degrees of freedom of optical signals.

On the other hand, the space–bandwidth product, which is the area of the smallest bounding perpendicular rectangle, may change significantly after linear canonical transformation and quadratic-phase optical systems. (This is the reason the number of samples must be increased at some intermediate stages of certain previously proposed FRT and LCT algorithms that rely on the

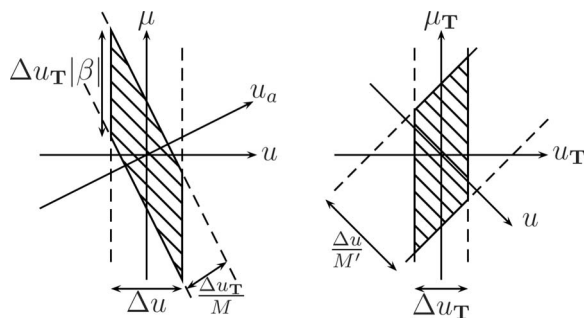


Fig. 4. Space–frequency support of  $f(u)$  (left) and  $f_T(u)$  (right) for space- and LCT-limited signals. The area of both parallelograms is equal to  $\Delta u \Delta u_T |\beta|$ .

space–bandwidth product either as the measure of the number of degrees of freedom or as the minimum number of samples required [69–75]. In contrast, fast computation of LCTs based on the results presented in [50] allows us to work with the same number of samples in both domains without requiring any oversampling. This number of samples is the minimum possible for both domains, and is given by the bicanonical width product.) This factor makes the conventional space–bandwidth product undesirable as a measure of the number of degrees of freedom, which we expect to be an intrinsic and conserved quantity under invertible unitary transformations.

The so-called *generalized space–bandwidth product*, which essentially removes the requirement that the rectangular support be perpendicular to the axes, has been proposed [81] as an improvement over the conventional space–bandwidth product. It has been noted that this entity is invariant under the FRT operation (rotational invariance), but it was emphasized that “further research is required in obtaining other forms of generalized space–bandwidth products that are invariant under more general area preserving time–frequency operations: the symplectic transforms” ([81], pp. 1238–1239). The bicanonical width product meets precisely this demand and allows us to compute LCTs with the minimum possible number of samples without requiring any interpolation or oversampling [50].

## 6. CONCLUSION

We showed that LCT domains correspond to scaled fractional Fourier transform (FRT) domains, and thus to scaled oblique axes in the space–frequency plane (phase space). Based on a many-to-one association of LCTs with FRTs, LCT domains can be labeled and monotonically ordered by the corresponding fractional order parameter instead of their usual three parameters, which do not directly lend themselves to a natural ordering. Despite their three parameters, LCT domains do not constitute a truly richer family of domains than FRT domains. LCTs of a signal that are associated with the same FRT order parameter essentially live in the same domain (the FRT domain with that order). We have referred to such LCT domains as essentially equivalent domains. If a signal is confined to a finite interval in a certain LCT domain, it will also be confined to a finite interval in all essentially equivalent domains.

An optical system consisting of arbitrary combinations of lenses, sections of free space, and quadratic graded-index media can be modeled as an LCT. The distribution of light at any plane perpendicular to the optical axis is given by an LCT of the input distribution. The equivalence of LCT domains to scaled FRT domains implies that the evolution of light through LCTs with different parameters can also be interpreted as an evolution through scaled FRT domains of monotonically increasing order. This offers a more transparent view of the propagation of light through the system, as opposed to imagining that the light goes through an unsequenced series of three-parameter domains whose whereabouts we cannot visualize.



The significance of the bicanonical width product, which is a generalization of the space–bandwidth product, can be established in a number of different ways. We used the LCT sampling theorem to conclude that the bicanonical width product is the minimum number of samples to uniquely identify a signal out of all possible signals whose energies are approximately confined to finite intervals in two given LCT domains. The bicanonical width product gives the number of degrees of freedom of this set of signals, and this number of samples can be used to reconstruct the signal by using the LCT sampling theorem.

Another line of development involves the equivalence between LCT and FRT domains. If one conventionally specifies the space–domain and frequency–domain extents for a set of signals, the space–frequency support becomes a rectangular region (Fig. 1). The space–bandwidth product is equal to the area of the rectangle. By using the fact that LCT domains correspond to scaled oblique axes in the space–frequency plane, we showed that when the extents are specified in two LCT domains, the space–frequency support becomes a parallelogram defined by the intersection of two oblique corridors (Fig. 5). The bicanonical width product is equal to the area of the parallelogram.

When confronted with a space–frequency support of arbitrary shape, it is quite common to assume the number of degrees of freedom to be equal to the space–bandwidth product, without regard to the shape of its space–frequency support. In reality, this is a worst-case approach that encloses the arbitrary shape within a rectangle perpendicular to the axes and overstates the number of degrees of freedom (see Fig. 5).

The bicanonical width product provides a tighter measure of the number of degrees of freedom than the conventional space–bandwidth product and allows us to represent and process the signals with a smaller number of samples, since it is possible to enclose the true space–frequency support more tightly with a parallelogram of our choice, as compared with a rectangle perpendicular to the axes, or indeed any rectangle. In some applications, where the underlying physics involves LCT type integrals (as is the case with many wave propagation problems and optical systems), parallelograms may be excellently, if not perfectly, tailored to the true space–frequency supports of the signals.

Another important feature of the bicanonical width product is that it is invariant under linear canonical

transformation. The fact that LCTs model an important family of optical systems makes the bicanonical width product a suitable invariant measure for the number of degrees of freedom of optical signals. On the other hand, the space–bandwidth product, which is the area of the smallest bounding perpendicular rectangle, may change significantly after linear canonical transformation.

Finally, we note that we can accurately compute an LCT with a minimum number of samples given by the bicanonical width product, so that the bicanonical width product is also a key parameter in fast discrete computation of LCTs [50].

Given the fundamental importance of the conventional space–bandwidth product in signal processing and information optics, we believe the bicanonical width product will also play an important role in these areas.

*Note added on proof:* We have recently been made aware of two new works [82,83] on the subject of fast LCT computation, which have a graphic illustration of the LCT algorithm mentioned in Subsection 2.B.5.

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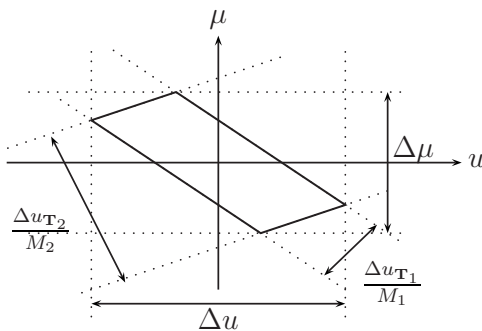


Fig. 5. Parallelogram shaped space–frequency support with area equal to the bicanonical width product  $\Delta u_{T1} \Delta u_{T2} |\beta_{1,2}|$ , which is smaller than the space–bandwidth product  $\Delta u \Delta \mu$ .

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